EEL 5544 Noise in Linear Systems Lecture 33

RANDOM PROCESSES (RPS)

Example: Noise in a Linear System

- Want to define RPs in very similar ways as RVs
- **Recall** A RV is a _____ that maps elements in Ω onto the real numbers

DEFN	A (real-valued) random process is a _	that maps elements in	Ω
	onto real-valued:		

- A RP can be denoted as X(t; ω). As for the case of random variables, we usually suppress
 the dependence of ω and just write X(t)
- The index of the function is usually taken to be t for time but can also represent space, etc.
- For a particular $\omega = \omega_0$, can draw $X(t; \omega_o)$
- $X(t; \omega_0)$ is a deterministic function and is called a _____ for $X(t; \omega)$ Example

_			
n		ית	N
	D .		
-			

A random process or stochastic process on a probability space (Ω, \mathcal{F}, P) is an ______ on (Ω, \mathcal{F}, P) .

- t typically denotes time with $\mathbf{T} \subset \mathbf{R}$
 - T can be discrete:

* $\mathbf{T} = \mathbf{Z}$ (the integers) * $\mathbf{T} = \{0, 1, 2, 3, \dots\}$

*
$$\mathbf{T} = \{0, 1, 2, 3, \dots\}$$

* $\mathbf{T} = \{\frac{1}{n}, \frac{2}{n}, \frac{3}{n}, \dots\}$

- or T can be continuous:
 - * $\mathbf{T} = \mathbf{R}$ * $\mathbf{T} = [0, \infty)$ * $\mathbf{T} = (a, b)$
- RPs can also be classified according to the values they take on:

DEFN	The <i>S</i> set of	of a RP X is the	(i.e., the
	S is a discrete set, then X is a _ S is a continuous set, then X is	a RP.	RP.

(The oscillator with random phase is an example of a continuous-time, continuous-amplitude random process.)

(The previous linear system example has inputs and outputs that are discrete-time, discrete-amplitude random process.)

Example

DISTRIBUTION AND DENSITY FUNCTIONS FOR RPS

- Recall that for each $t \in \mathbf{T}$, X(t) is a RV
- Thus for each $t \in \mathbf{T}$, $\{X(t) \leq x\} \in \mathcal{A}$ (is an event)

$$\Rightarrow \bigcap_{k=1}^{n} \{ X_{t_k} \le x_k \} \in \mathcal{A}$$
$$\Rightarrow P\left[\bigcap_{k=1}^{n} \{ X_{t_k} \le x_k \}\right]$$

is defined.

 \Rightarrow We can define distribution functions for the random process

Let X(t) be a random process on a probability space (Ω, \mathcal{F}, P) .

DEFN The *one-dimensional cumulative distribution function* for X(t) is given by

DEFN The *n*-dimensional cumulative distribution function for X(t) is given by

$$F_{X,n}(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n) = P [X (t_1) \le x_1, X (t_2) \le x_2, \dots, X (t_n) \le x_n] = P \left[\bigcap_{k=1}^n \{X(t_k) \le x_k\} \right]$$

– For our purposes, a RP is completely specified by its n-dimensional cdfs for all positive integers n

DEFN

The *n*-dimensional probability density function for a random process X(t) is the function $f_{X,n}(x_1, x_2, ..., x_n; t_1, t_2, ..., t_n)$ such that

$$F_{X,n}(x_1, x_2, ..., x_n; t_1, t_2, ..., t_n) = \int_{-\infty}^{x_1} \cdots \int_{-\infty}^{x_n} f_{X,n}(u_1, ..., u_n; t_1, ..., t_n) du_n ... du_1.$$

DEFN The *n*-dimensional probability mass function for a discrete-valued random process X(t) is given by

$$p_{X,n}(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n)$$

= $P[X(t_1) = x_1, X(t_2) = x_2, \dots, X(t_n) = x_n]$

- Let the state space for X(t) be given by

$$\mathcal{S} = \left\{ s_k \, | k \in \mathbf{N} \right\},\,$$

where $s_i < s_j$ whenever i < j.

- Then

$$p_{X,1}(s_i;t) = F_{X,1}(s_i;t) - F_{X,1}(s_{i-1};t),$$

and

$$p_{X,2}(s_i, s_j; t_1, t_2) = F_{X,2}(s_i, s_j; t_1, t_2) - F_{X,2}(s_{i-1}, s_j; t_1, t_2) - F_{X,2}(s_i, s_{j-1}; t_1, t_2) + F_{X,2}(s_{i-1}, s_{j-1}; t_1, t_2)$$

Example